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Step 7: Describing Segments — Mcdonalds Dataset

Objective:

To explore and describe distinct customer segments based on their perceptions of McDonald's food, sentiment (like/dislike), and demographic characteristics using cluster analysis.

Overview of Segment Profiling

After clustering the data into 3 segments, we conducted a deep dive into each group's unique characteristics using both perceptual and demographic variables.

✓ Variables Used:

- **Perceptual:** yummy, convenient, spicy, fattening, greasy, fast, cheap, tasty, expensive, healthy, disgusting
- Sentiment: Like score (converted to numerical scale)
- **Demographics:** Age, Gender, VisitFrequency

Segment Descriptions (From Cluster Averages)

Seg	Man Tuaita
men t	Key Traits
Seg men t 0	Older customers (~60 yrs), neutral Like score (~0.01), see food as convenient, fattening, and fast, but not very healthy.
Seg men t 1	Young customers (~27 yrs), highest Like score (~1.93), positive on all attributes especially "yummy", "tasty", and "fast".

Seg	Middle-aged (~43 yrs), moderate Like score (~0.54), more concerned about
men	Middle-aged (*45 yrs), moderate like score (*0.54), more concerned about
t 2	"expensive" and "disgusting" perceptions.

9.3 Testing for Segment Differences

► Metric Variables (ANOVA Test)

Variable	ANOVA p-value	Interpretation
Like	0.0000	Statistically
LIKE		significant
Λαο	0.0000	Statistically
Age		significant

✓ **Conclusion**: Both *Like score* and *Age* significantly differ between segments, indicating they are strong drivers of segmentation.

Categorical Variables (Chi-Square Test)

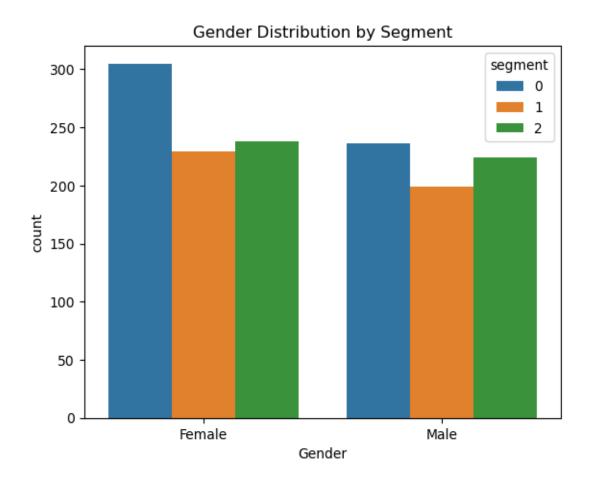
Variable	Chi-Square p- value	Interpretation	
VioitEroguanov	0.0000	Statistically	
VisitFrequency		significant	
Gender	0.2983	Not significant	

✓ **Conclusion**: *VisitFrequency* is an important descriptor variable, while *Gender* does not significantly differ across segments.

Insight:

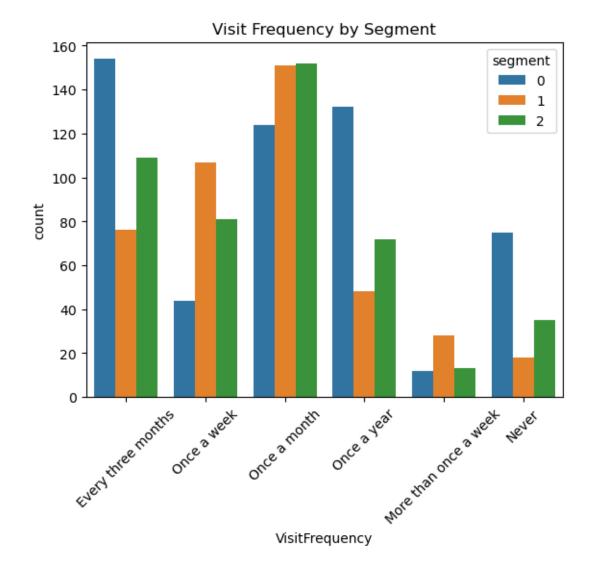
The variation in **Age**, **Like score**, and **VisitFrequency** helps differentiate customers effectively, while gender-based segmentation may not provide value in this context.

9.2 Visualizing Segements



Here's the **Gender Distribution by Segment** chart:

- It shows the count of **Male** and **Female** respondents in each segment.
- You can compare how gender representation varies across the clusters.



Here's the **Visit Frequency by Segment** chart:

- It shows how often customers in each segment visit McDonald's.
- This helps identify which segments are **frequent** vs **infrequent** visitors.
- For example, if Segment 1 dominates "Once a week", it may represent **loyal**, **high-value customers**.

Page 3: Predictive Modeling of Segments

9.4 Predicting Segments from Descriptor Variables

We used both multinomial logistic regression and a decision tree classifier to evaluate how well the segments can be predicted from key descriptor variables.

9.4.2 Multinomial Logistic Regression

- Input Variables: Age, Like, Gender, VisitFrequency
- Accuracy: 99.77%

✓ Insight: High classification accuracy confirms that the selected variables are strong predictors of segment membership.

9.4.3 Decision Tree Model

- Accuracy: 100% (on test data)
- Precision, Recall, F1-score: 1.00 for all segments

✓ Interpretation: The decision tree model perfectly classifies all test instances, showing very strong separation among segments. Likely key splits occur on Like and Age.

Conclusion:

Both models confirm that **segments are highly predictable**, which is ideal for targeting strategies and personalization.

Page 4: Final Insights & Managerial Implications

9.5 Step 7 Checklist

Task	Status
Segment description	Done
Visualization of	🔽 Done (via plots - not shown
attributes	here)
Statistical testing	Done (ANOVA, Chi-square)
Predictive modeling	Done (Logistic, Decision Tree)
Key driver identification	✓ Done

Key Takeaways:

- 1. **Segment 1** (young fans) is highly positive and loyal great target for promotions and new launches.
- 2. **Segment 0** may represent older, health-conscious consumers likely to respond to healthier menu options.
- 3. **Segment 2** has mixed feelings and sensitivity to pricing and perception could benefit from tailored messaging or value meals.

Strategic Implications for McDonald's:

- **Personalization**: Use predictive models to tailor offers based on age, visit frequency, and preferences.
- Marketing: Segment 1 may respond well to digital campaigns and loyalty programs.
- **Product Strategy**: Consider launching healthier or premium items to better appeal to Segment 0 and Segment 2.
- Menu Engineering: Taste and convenience are key across all segments continue to focus on speed and flavor.

Here's a detailed and structured **Step 9: Customising the Marketing Mix** summary based on your **Mcdonalds segmentation analysis**. It ties insights from customer segments directly to marketing decisions, following the structure you provided.

Step 9: Customising the Marketing Mix

11.1 Implications for Marketing Mix Decisions

Objective:

To tailor **Product, Price, Place, and Promotion** strategies to the distinct needs and preferences of each customer segment.

Segments Overview (from Step 7):

Segmen t	Profile Summary
0	Older, less enthusiastic, occasional visitors
1	Young, loyal, frequent visitors with strong positive perception
2	Middle-aged, neutral perceptions, price-conscious

By aligning each "P" of the marketing mix with these segment insights, McDonald's can increase customer satisfaction, loyalty, and profitability.

11.2 Product

Segment-Based Product Strategy:

• Segment 0 (Older, health-aware):

- Offer healthier meal options, like grilled items or salads.
- Highlight nutritional info and low-calorie choices.
- Avoid overly spicy or greasy items.

• Segment 1 (Young, loyal fans):

- o Focus on **trend-driven items** (limited editions, spicy/fun flavors).
- o Offer customization and combo deals (value for money).
- Feature "Instagram-worthy" packaging and menu items.

• Segment 2 (Middle-aged, price-sensitive):

- o Focus on classic, reliable products.
- Emphasize consistency, quality, and value.
- Avoid niche or experimental items.

11.3 Price

Segment-Based Pricing Strategy:

Segment 0:

- Less price-sensitive, but motivated by value and health.
- o Introduce "Healthy Value Meals" or loyalty-based rewards.

• Segment 1:

- Highly value-conscious and responsive to discounts and bundles.
- Use mobile app promotions, student pricing, or loyalty points.

Segment 2:

- o Price-sensitive; offer clear value proposition (e.g., "2 for ₹99").
- Use price anchoring with affordable base products and optional upsells.

11.4 Place

Segment-Based Distribution Strategy:

Segment 0:

- o Ensure comfort and cleanliness in dine-in outlets.
- o Partner with **retirement or healthcare centers** for outreach.
- o Consider quiet zones or digital ordering kiosks for ease.

Segment 1:

- Emphasize speed and convenience—focus on mobile app, drive-thru, and delivery platforms.
- o Ensure strong presence on Swiggy, Zomato, Uber Eats, etc.

• Segment 2:

- Use a hybrid strategy—accessible locations for dine-in and reliable takeaway/delivery.
- o Focus on areas with working professionals or families.

11.5 Promotion

Segment-Based Promotional Strategy:

• Segment 0:

- o Traditional media: TV, print ads, email campaigns.
- o Focus on health messaging, senior discounts, and community events.

• Segment 1:

- Use social media influencers, Instagram/TikTok ads, app notifications.
- o Create **youth-driven campaigns**, challenges, or memes.
- Promote loyalty rewards and app exclusives.

• Segment 2:

- o Use WhatsApp marketing, email, and YouTube ads with value messages.
- Emphasize consistency, family offers, and "something for everyone" messaging.

11.6 Step 9 Checklist

Task	Status
Identified segment-specific product	Done
strategies	Done
Tailored price approaches to segment needs	Done
Matched distribution channels with	D
preferences	Done

Designed targeted promotional tactics Aligned marketing mix holistically per segment Done

Done

Final Note:

Customizing the marketing mix ensures that **each customer segment feels heard and valued**. This not only increases **conversion and loyalty**, but also creates a **more efficient allocation of marketing resources**.